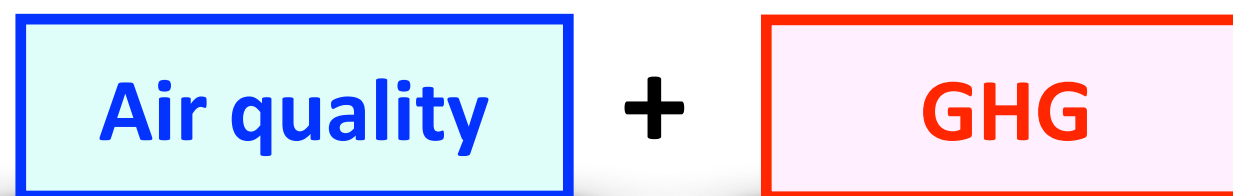
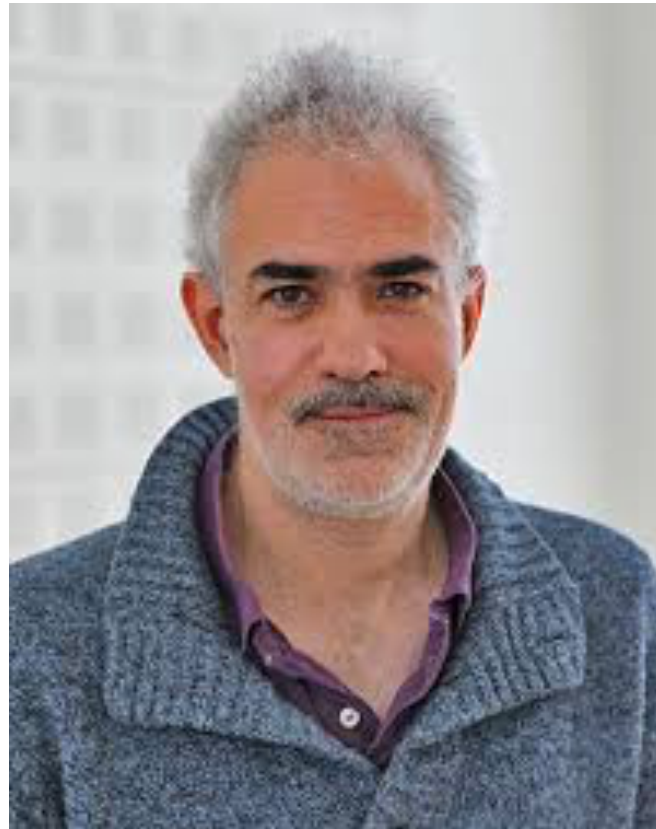


Predicting FF CO₂ fluxes using top-down NO_x and CO emissions estimated from multi-constituent chemical data assimilation

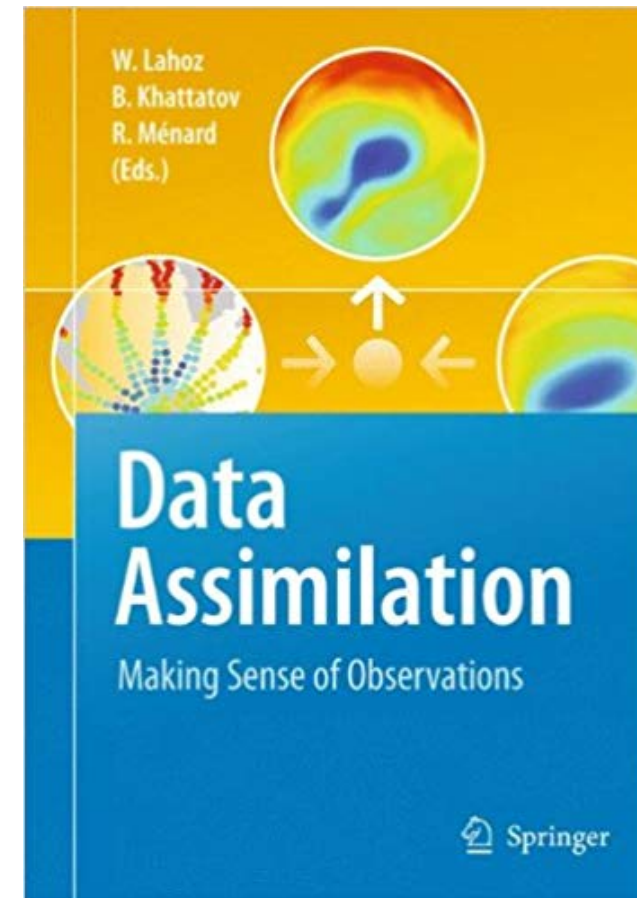


Kazuyuki Miyazaki, Kevin Bowman

Jet Propulsion Laboratory, California Institute of Technology



Dr. William Lahoz
1960~2019





Edo Bridge
東都江戸橋日本橋 (歌川広重)

GHG

- OH coupling (CH_4)
- Combustion process (NO_x , CO , CO_2)
- Joint emission optimization

Reuter et al., 2019

- The use of proxy species (NO_2 , CO) for CO_2 flux estimates: Contain a strong signal associated with human activities

Silva and Arellano,
2017, Tang et al., 2019

- AQ-GHG emission ratios can be used to understand emission processes (combustion type, new technology and regulation) and improve bottom-up inventories

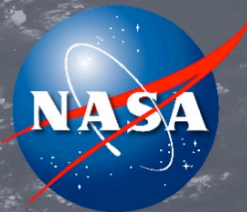
Konovalov et al., 2016

- Emission ratios can be used in hybrid emission estimations (e.g., from top-down NO_x to CO_2)

Multi-constituent chemical reanalyses

+

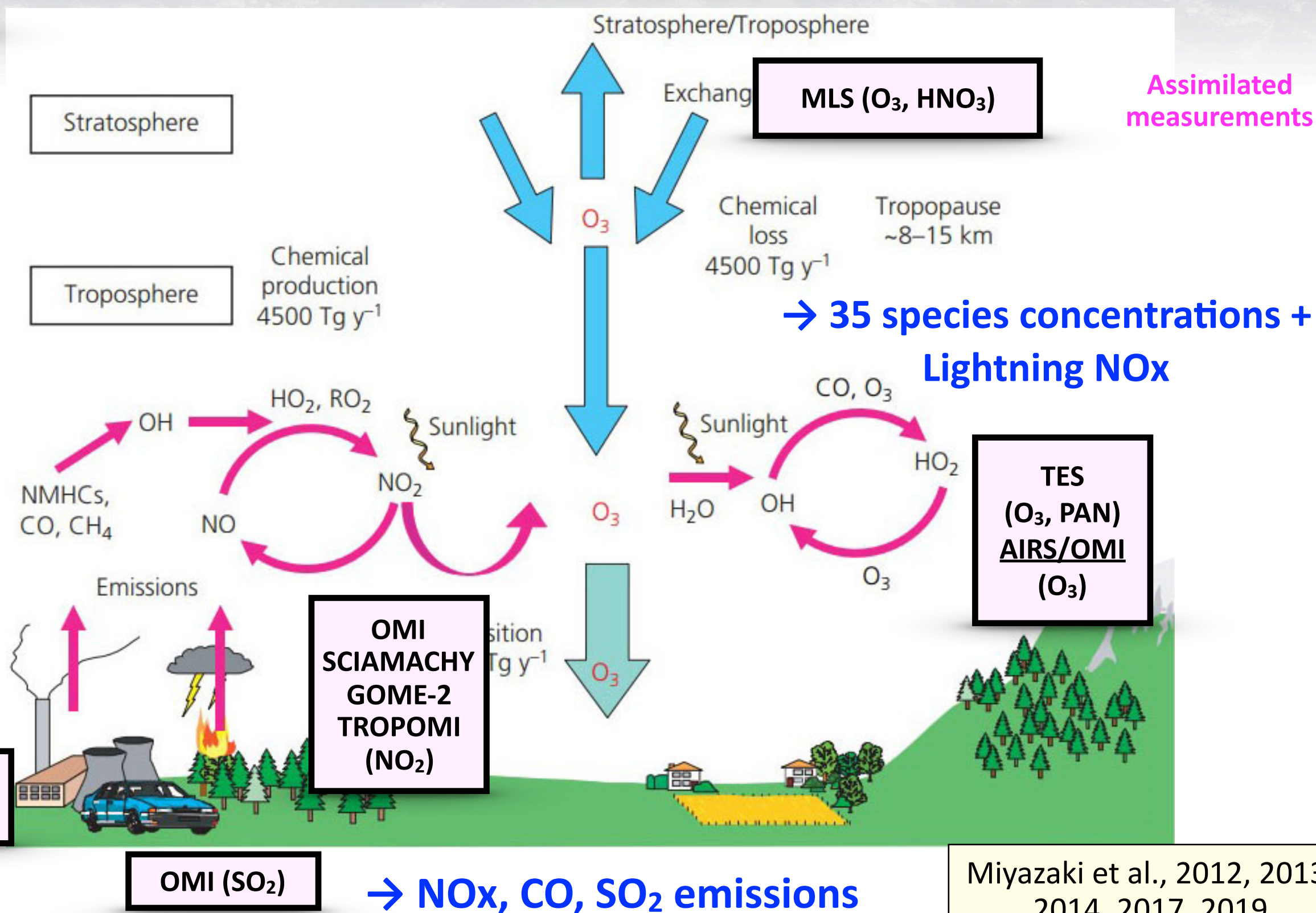
GHG inventories



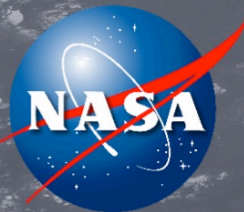
Multi-constituent chemical data assimilation

AQ

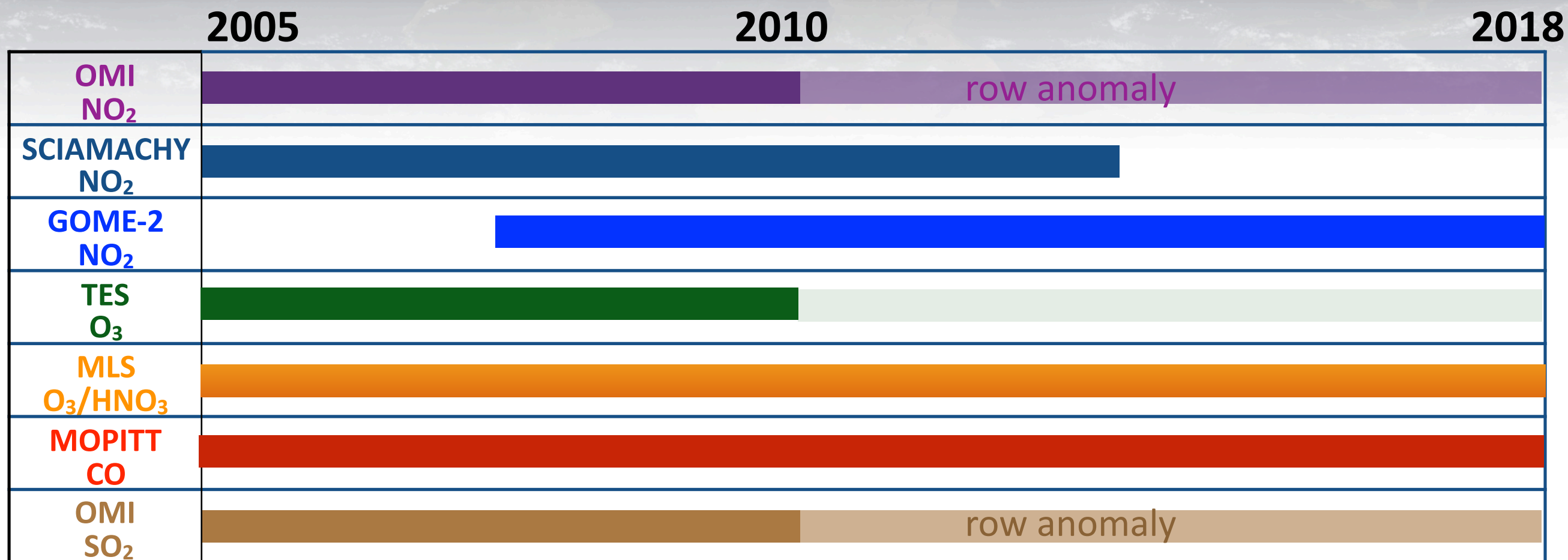
through ingestion of a suite of measurements from multiple satellite sensors



Miyazaki et al., 2012, 2013, 2014, 2017, 2019

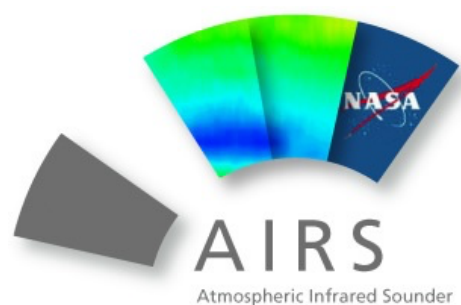


Tropospheric chemistry reanalysis (TCR-2)



*Two-hourly,
1.1°x1.1° resolution,
up to 70 hPa level*

- (1) understand the processes controlling the atmospheric environment
- (2) provide initial/boundary conditions for climate/chemical simulations
- (3) evaluate climate models and bottom-up emission inventories
- (4) suggest developments of models/observations (e.g., satellite concepts)



O₃ (AIRS/OMI)



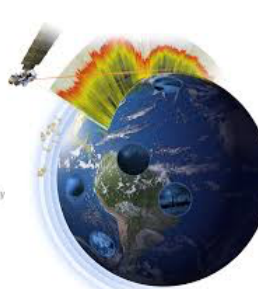
NO₂ (QA4ECV)
SO₂ (NASA PCA)



NO₂ (QA4ECV)



NO₂ (QA4ECV)



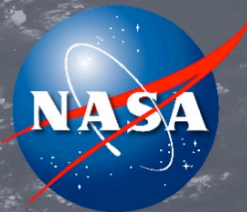
O₃ (v6)



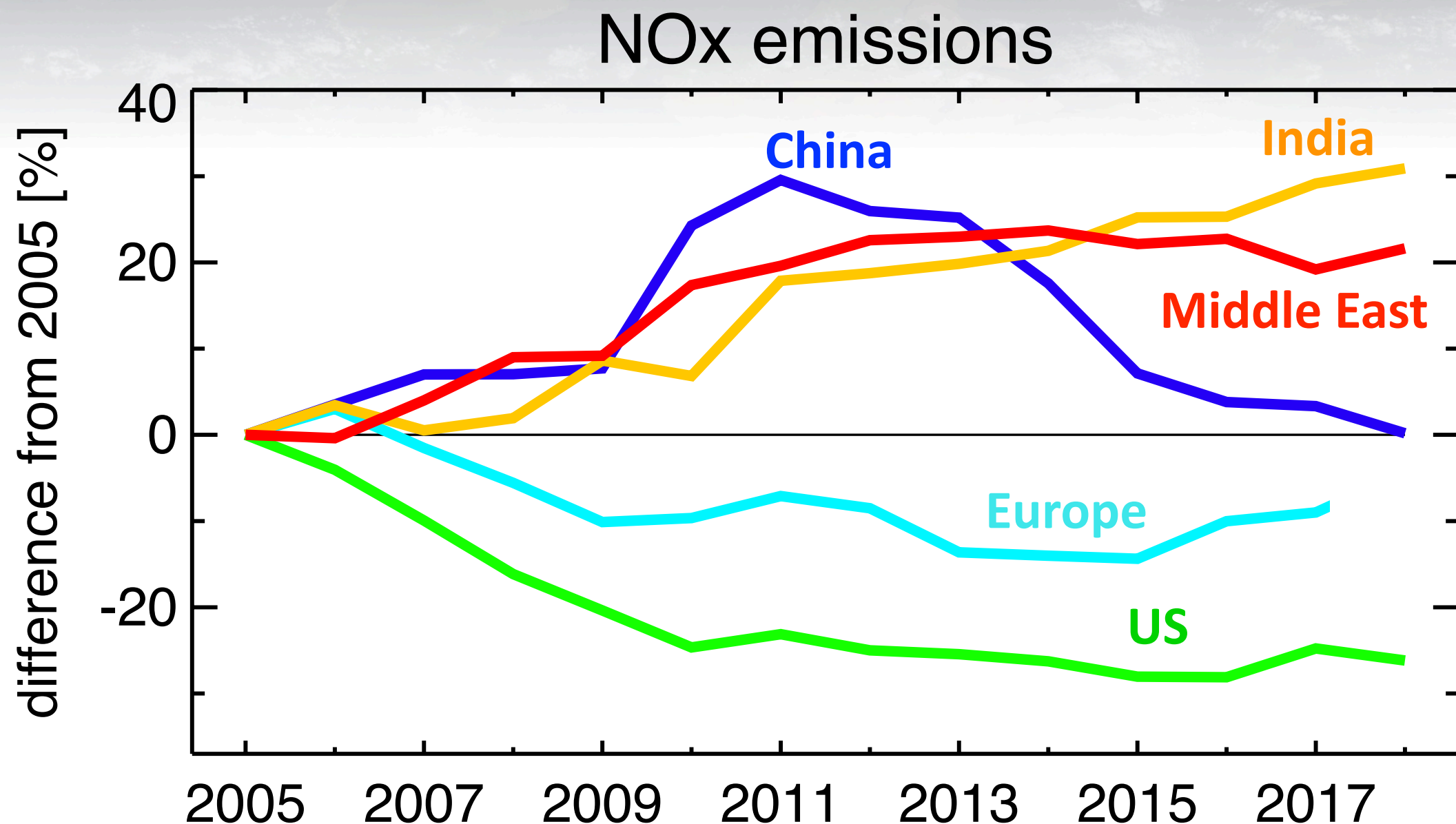
CO (v7J)



O₃, HNO₃ (v4.2)

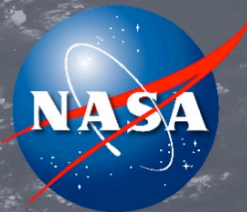


Global NO_x emission trends (2005-2018)

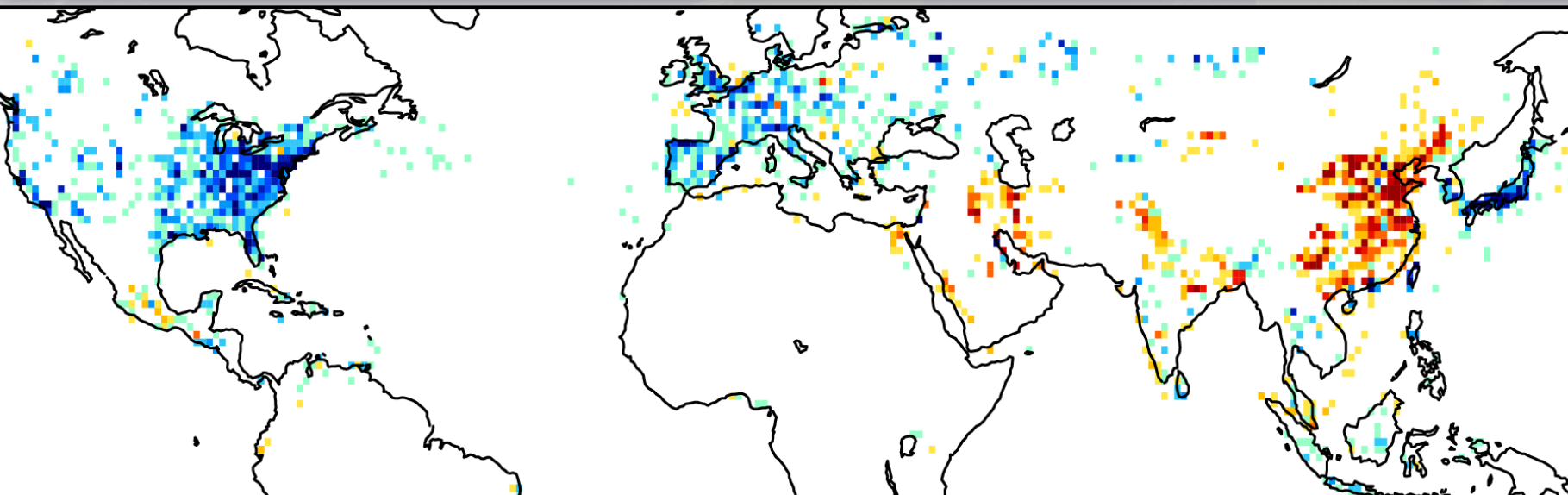


TCR-2 performance has been evaluated using various independent data (Miyazaki et al., in prep)

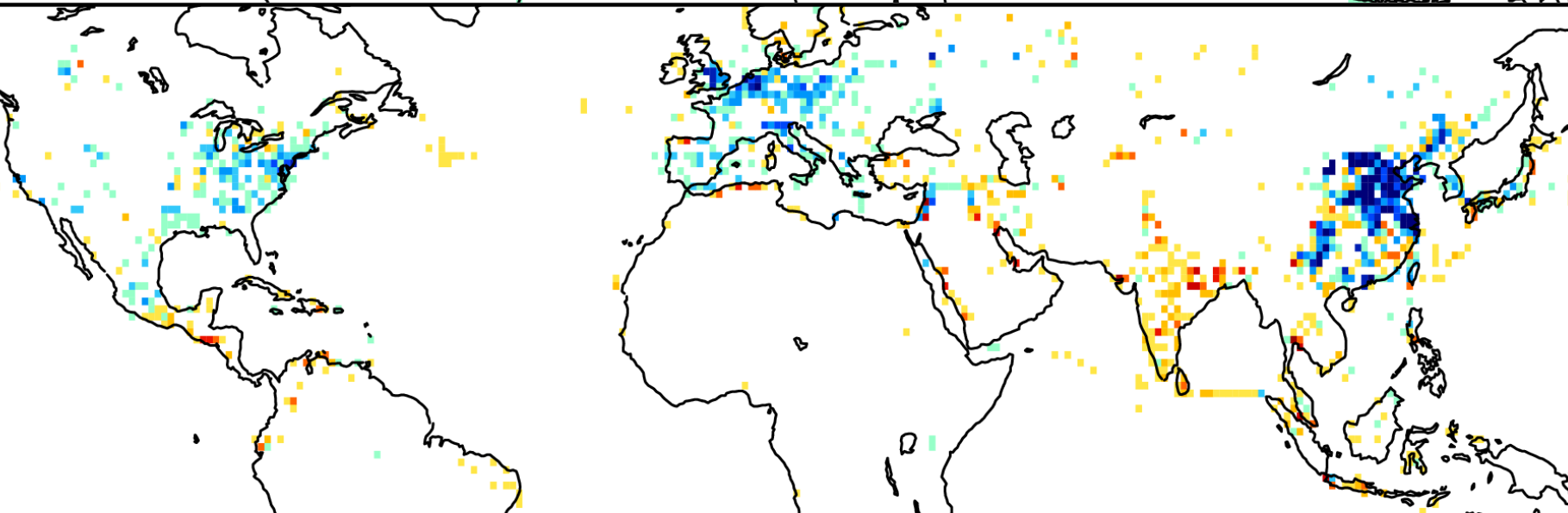




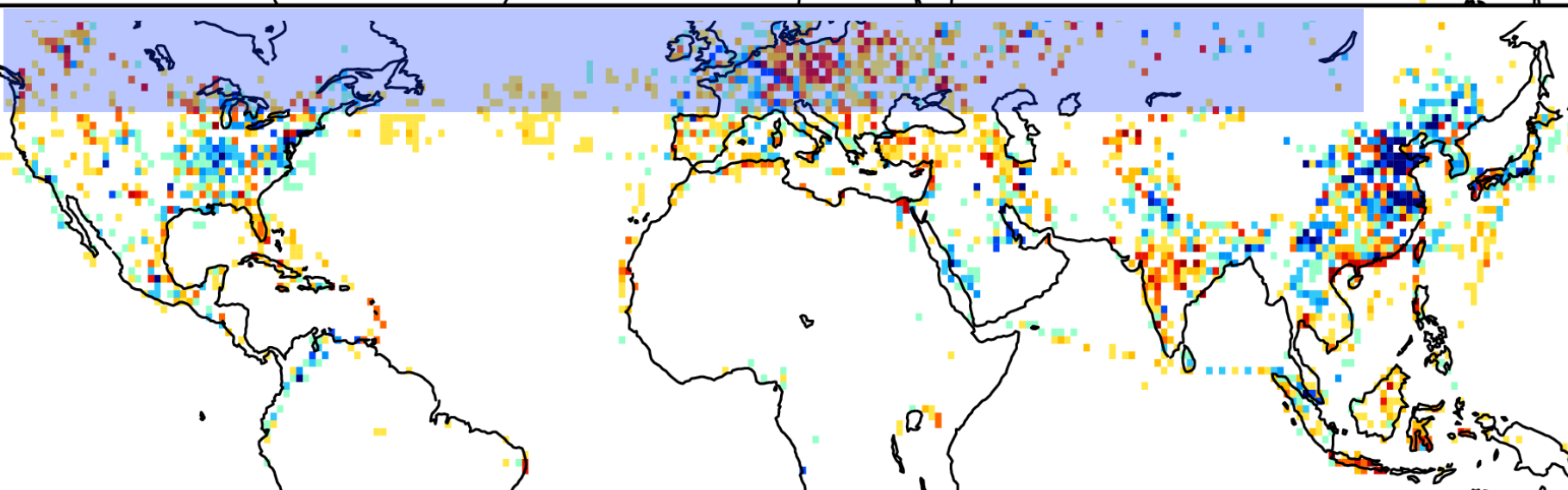
Global NO_x emission trends (2005-2018)



2005-2010

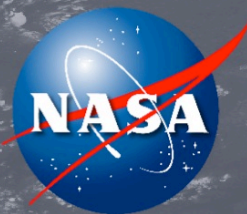


2010-2015



Insufficient constraints at NH high latitudes for 2017-2018 (unhealthy OMI only). Will be revised using OMI+GOME-2.

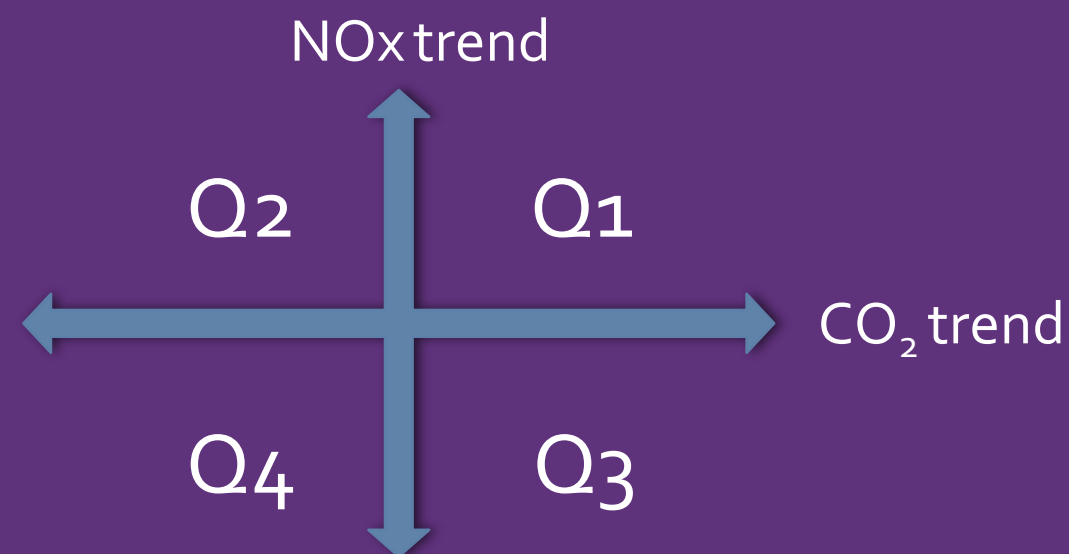
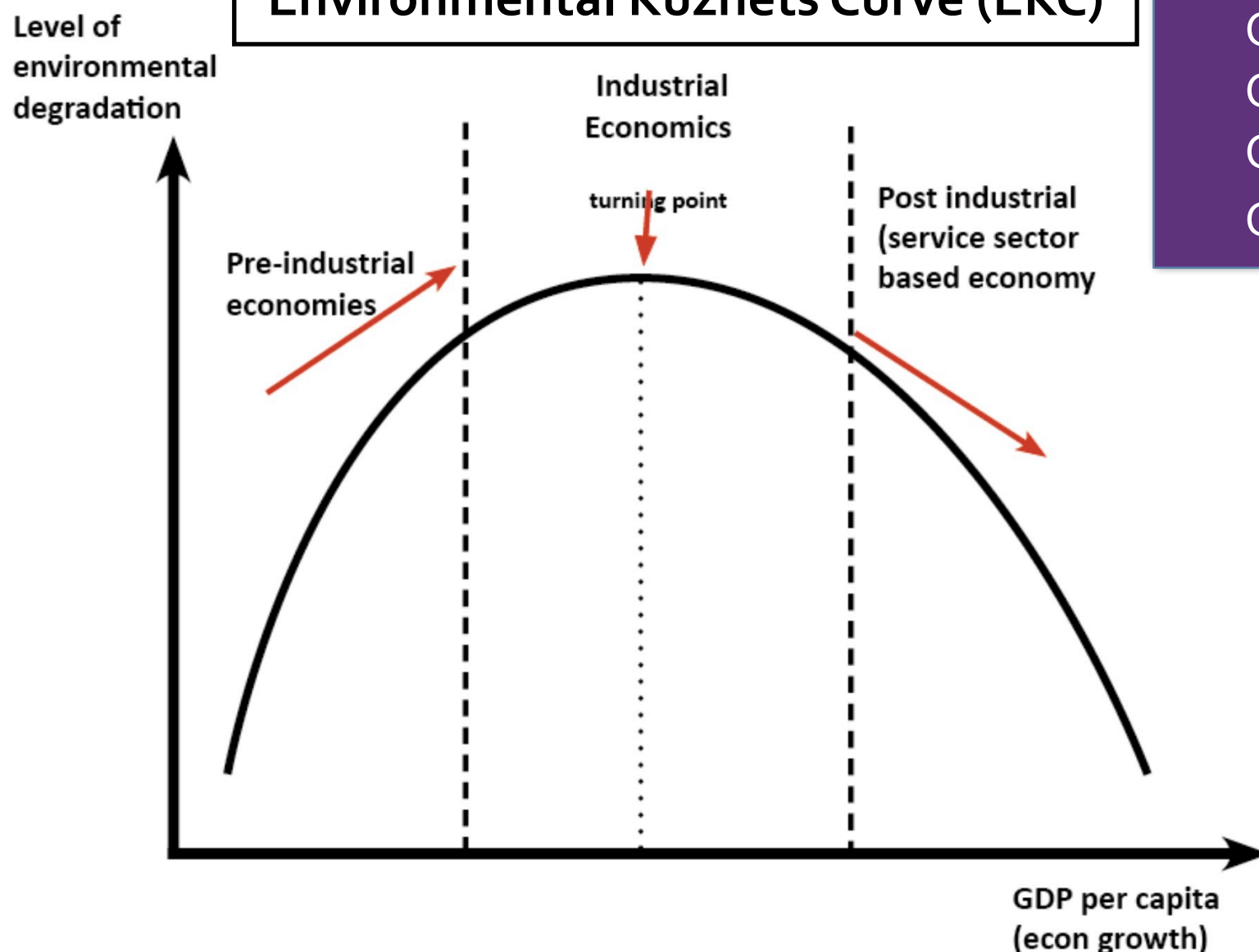
2015-2018



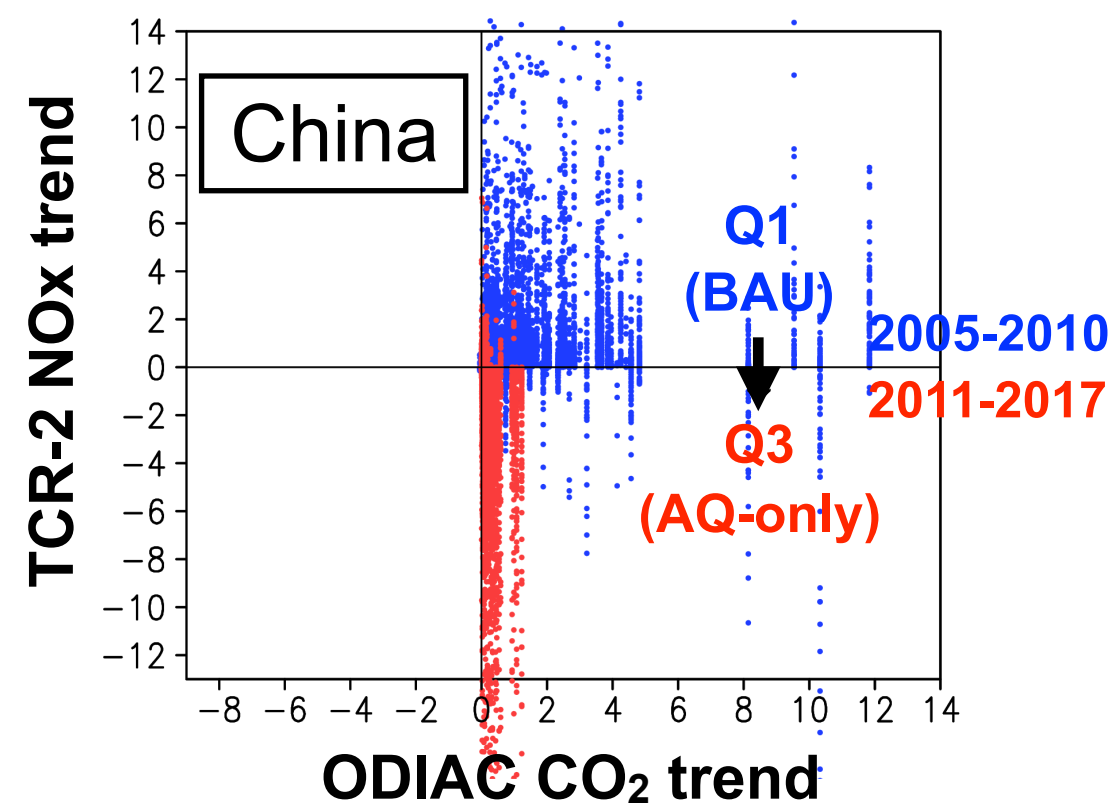
AQ/Carbon co-evolution

How will changes in air quality mitigation impact carbon emissions?

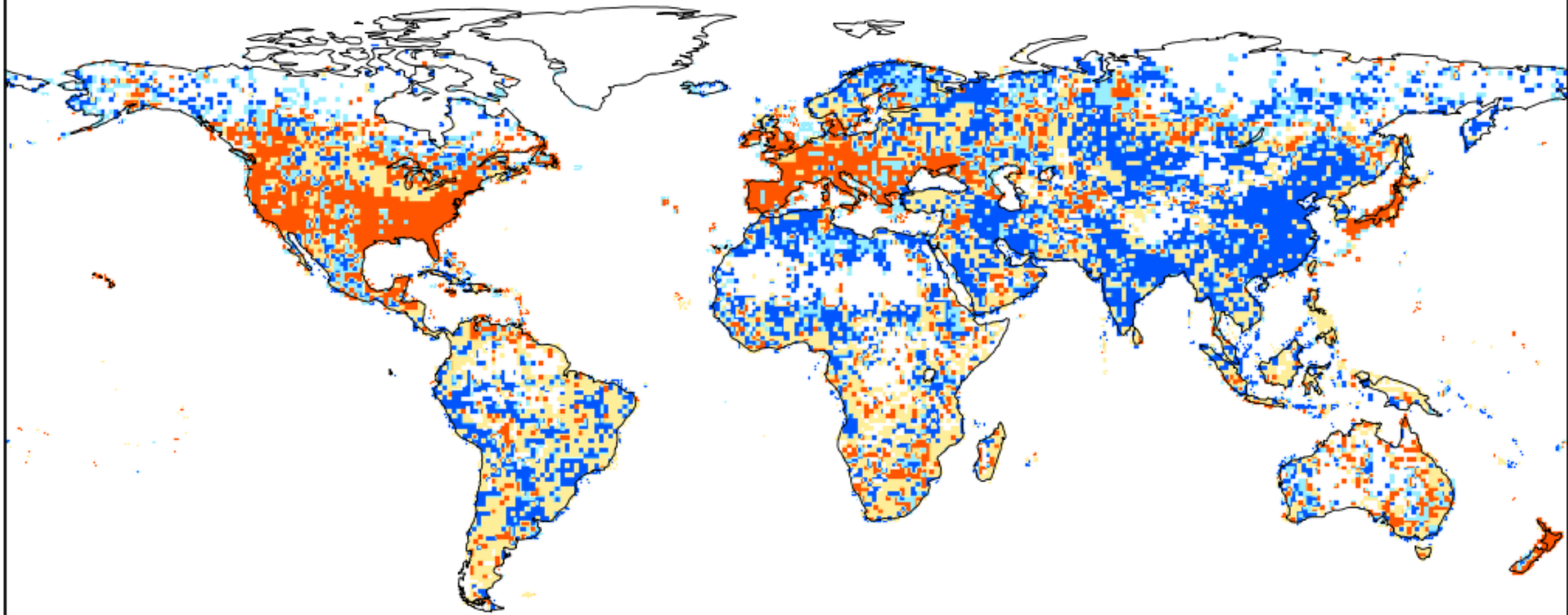
Environmental Kuznets Curve (EKC)



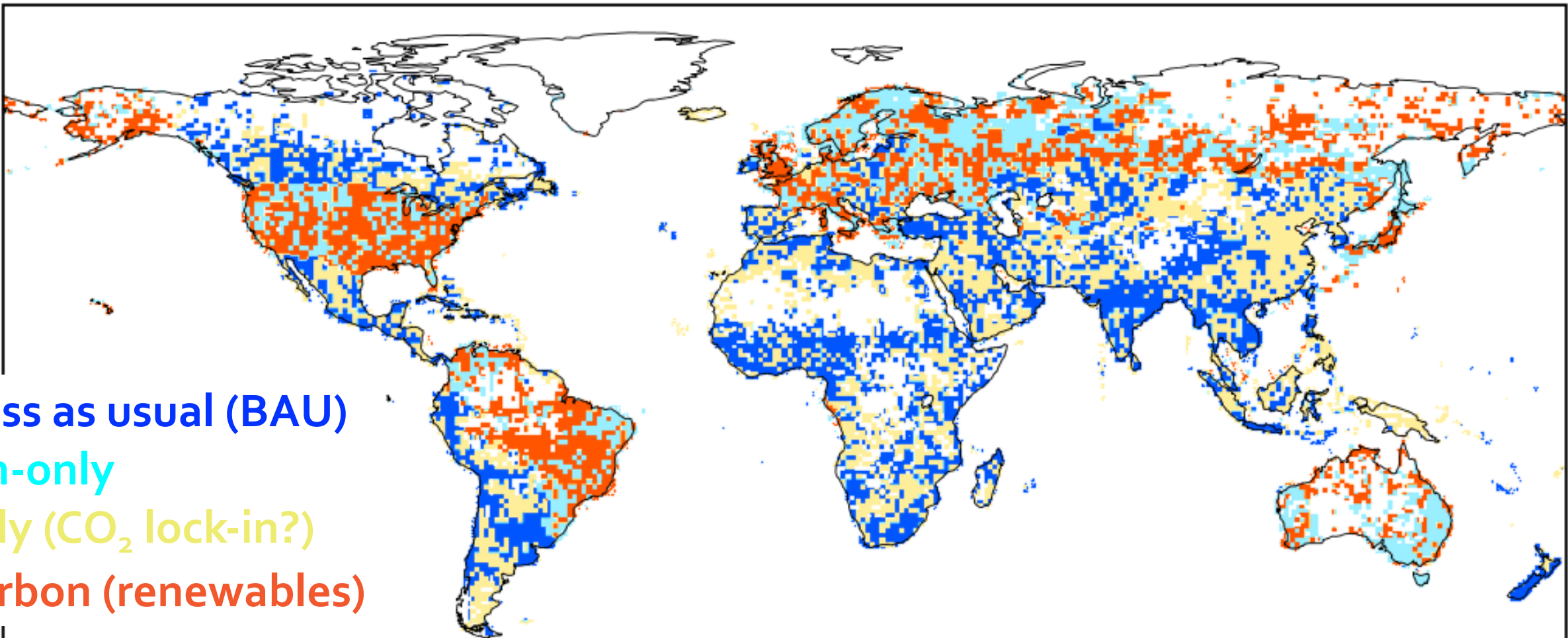
Q1: Business as usual (BAU)
Q2: Carbon-only
Q3: AQ-only (CO₂ lock-in?)
Q4: AQ/Carbon co-reduction (renewables)



**2005-
2010**



**2011-
2017**

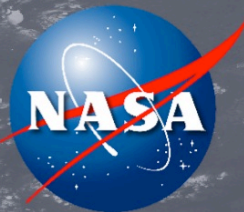


Q1: Business as usual (BAU)

Q2: Carbon-only

Q3: AQ-only (CO₂ lock-in?)

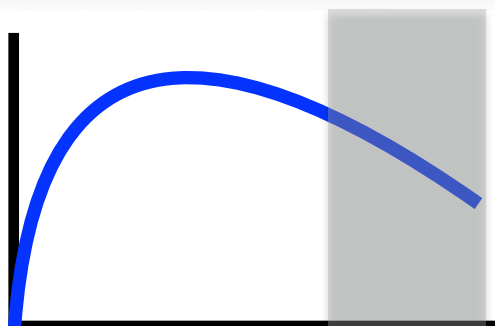
Q4: AQ/Carbon (renewables)



CO₂ flux prediction using top-down NO_x emissions

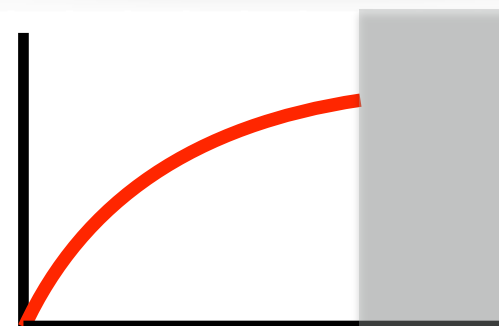
Air quality (NO_x)

Top-Down
quick update



GHG (CO₂)

Bottom-up
slower update

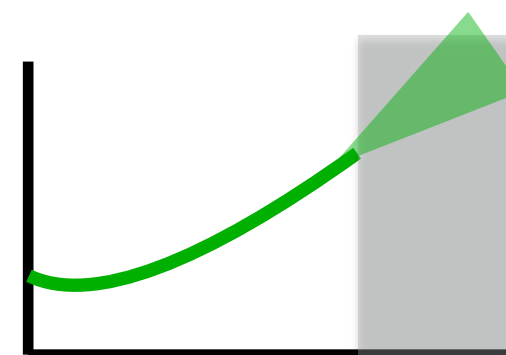


→ Time

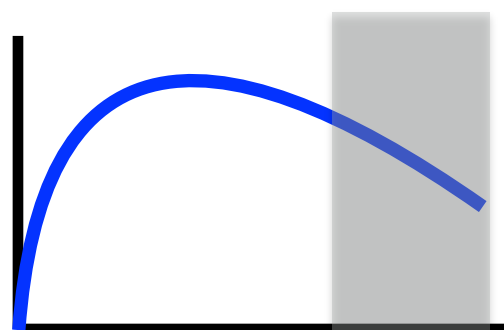
Variations in emission ratios (CO₂/NO_x)

(gradual changes in technology and regulation)

+Kalman filter prediction and error estimation

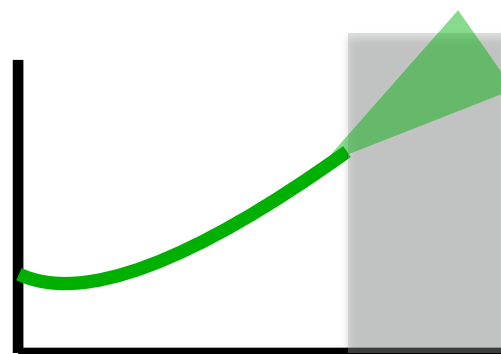


Top-Down NO_x emission



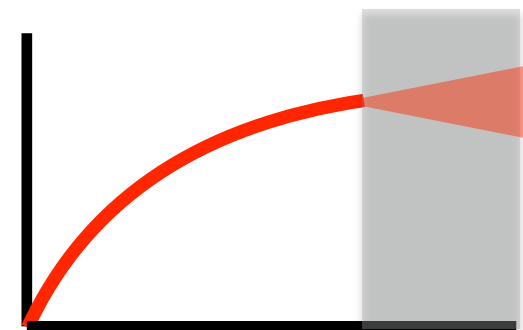
x

Emission ratio

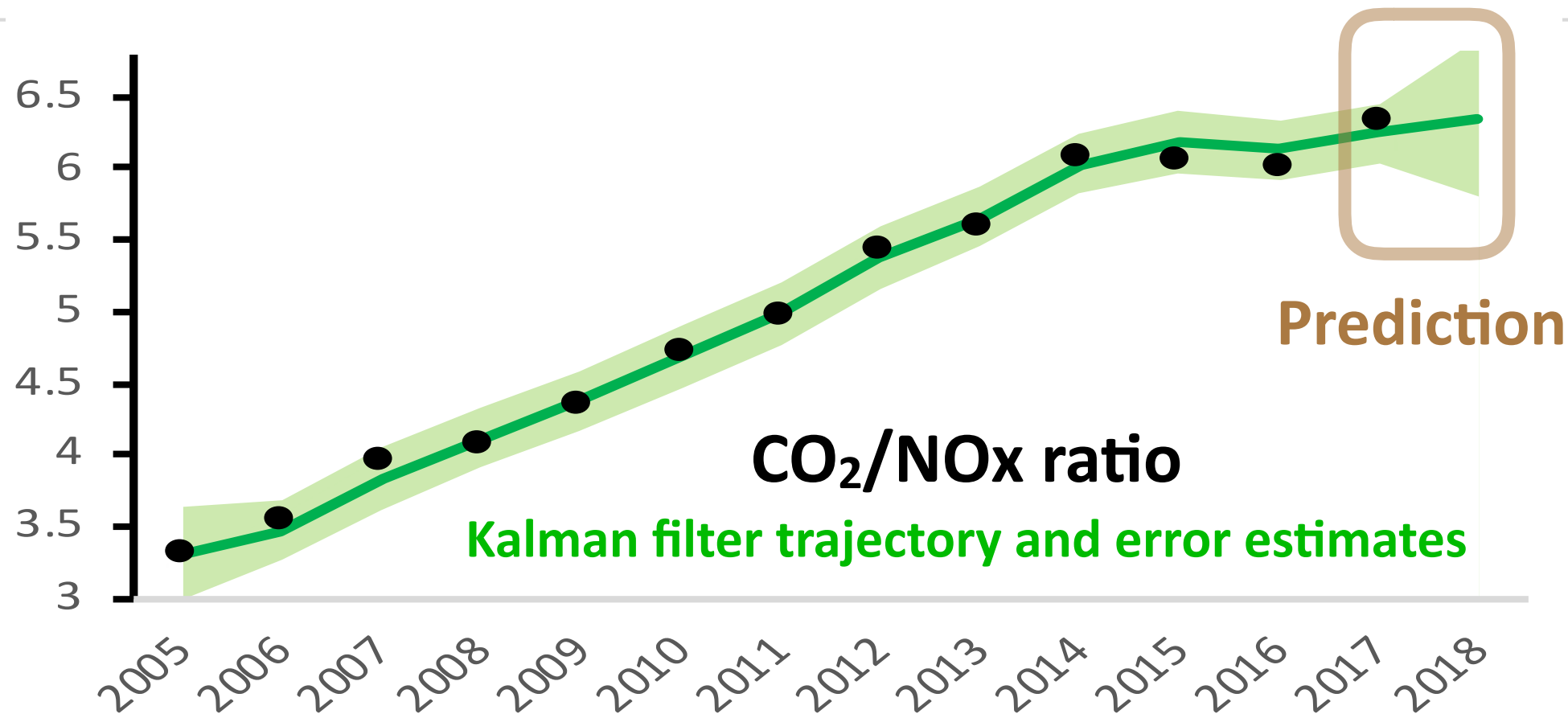
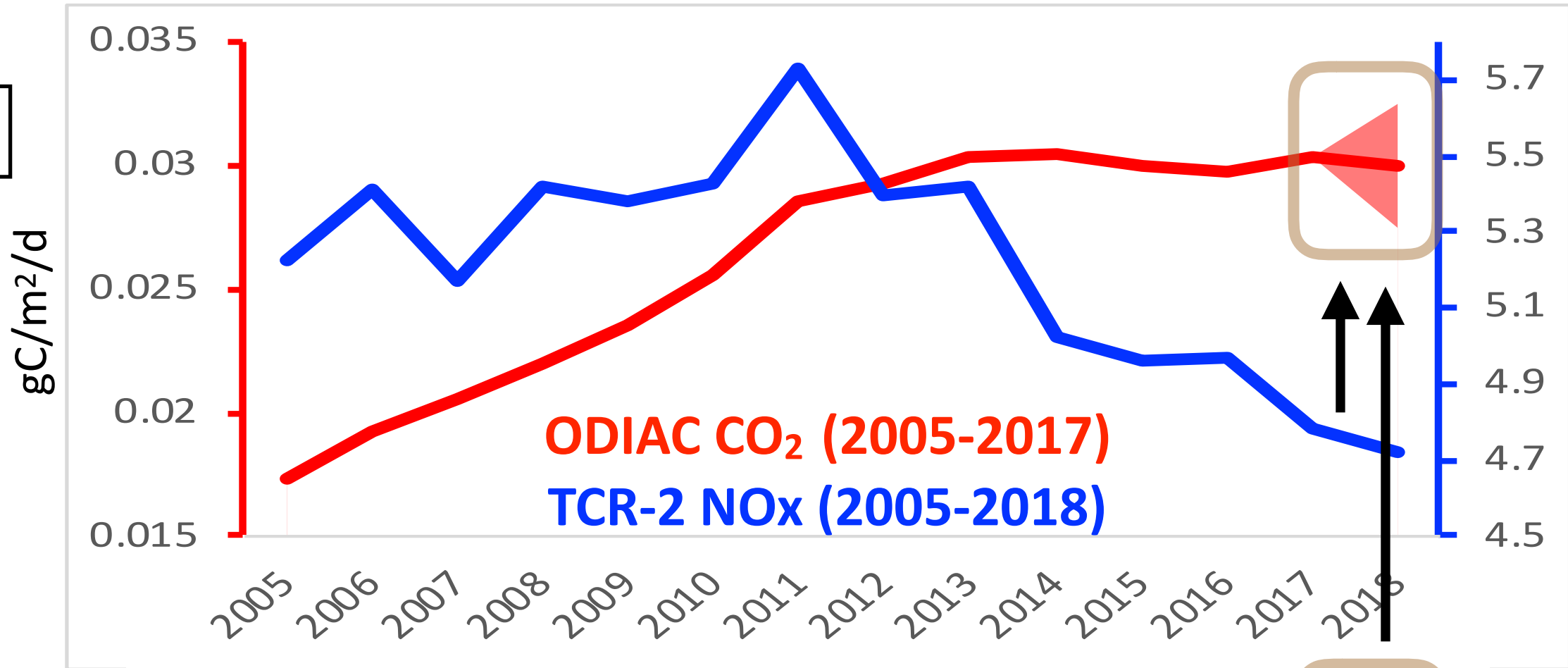


=

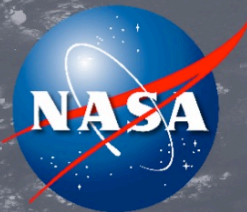
CO₂ flux prediction



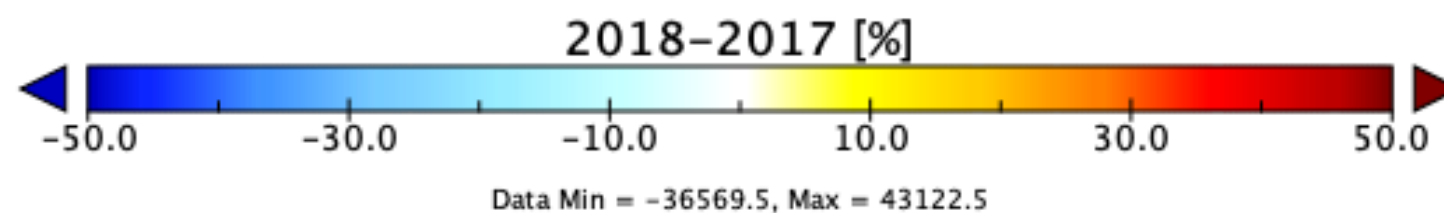
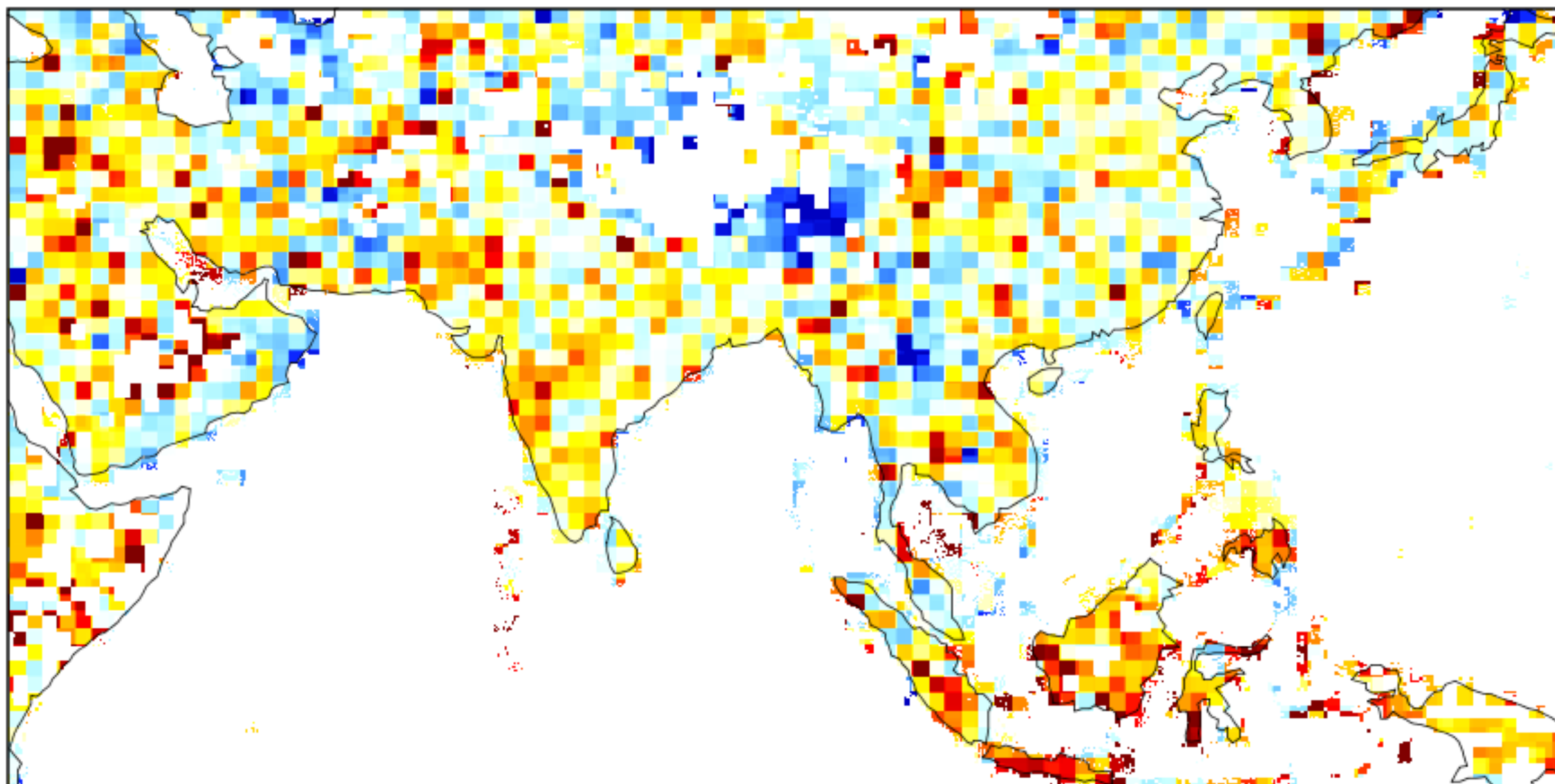
Shanghai



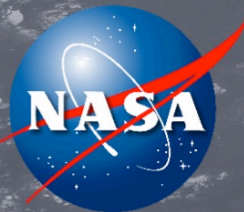
*Strong variations in emission factors for India, SE Asia, and the Middle east
→ Evaluation of emission inventories, understanding of emission processes*



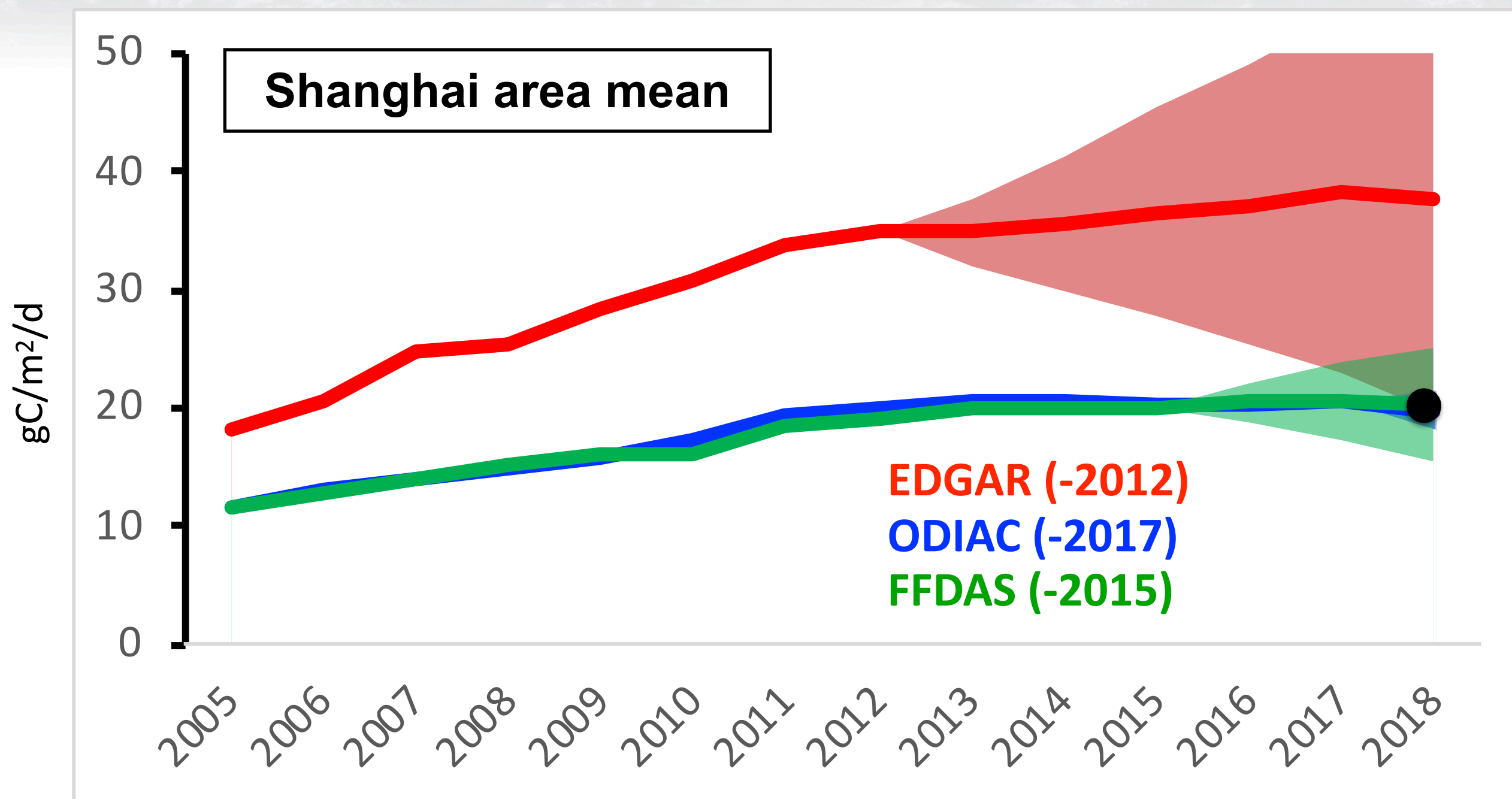
FF CO2 fluxes: 2018(predicted)-2017 (ODIAC)



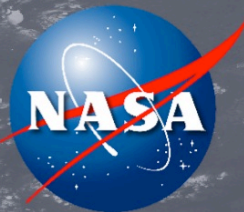
How will ODIAC 2018 look like? 😊



CO₂ flux estimations: Multi-inventories integration

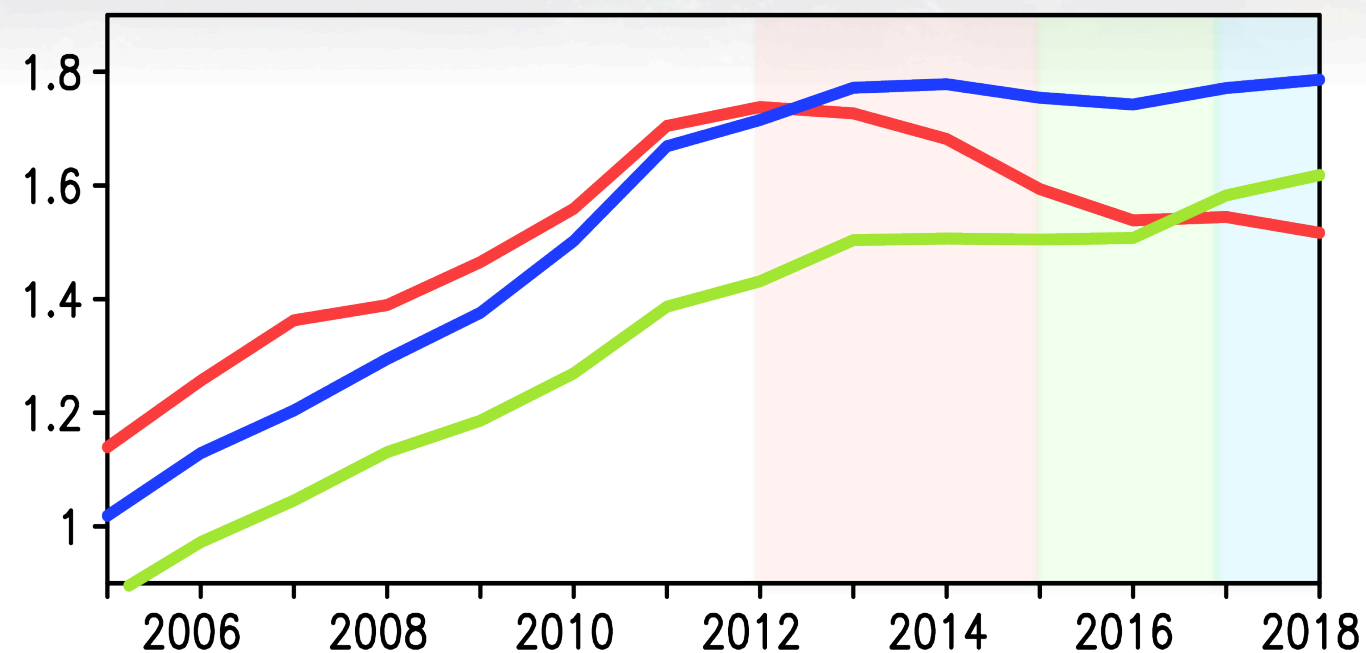


Multi-inventories mean (uncertainty-weighted) for 2018 : 19.9±1.5 gC/m²/d

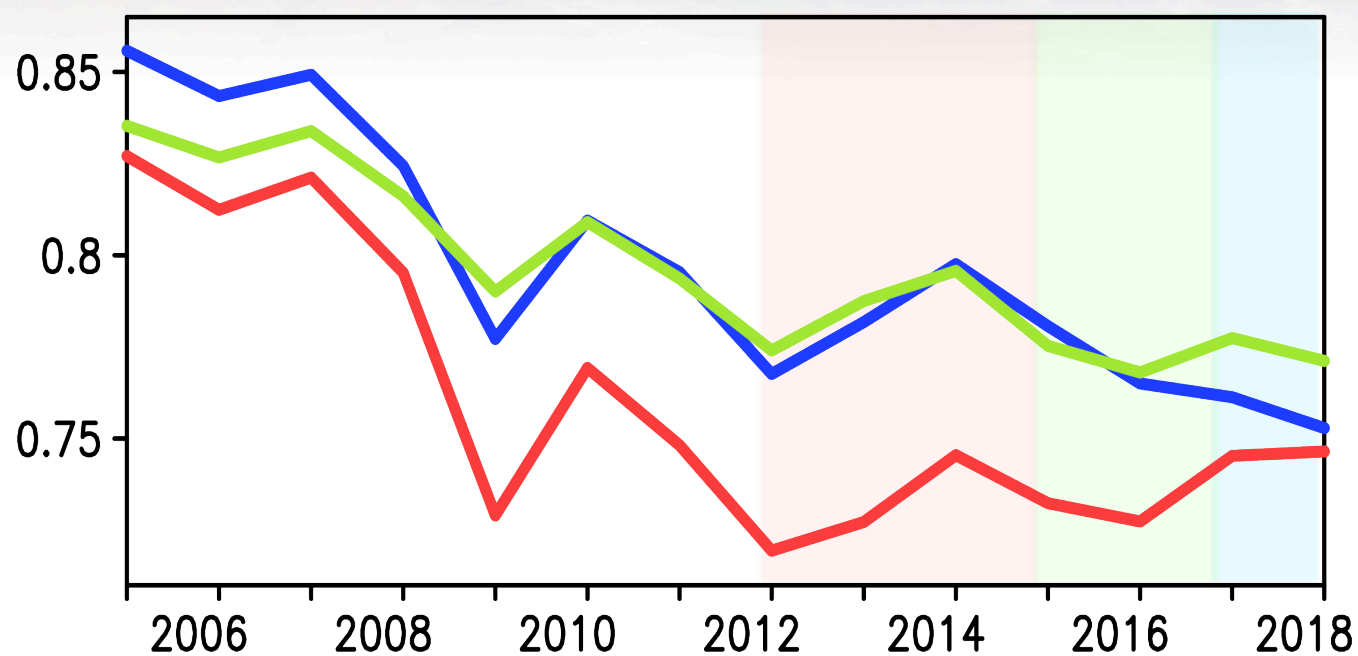


ODIAC 2005-2017 & predicted 2018 CO₂ fluxes (PgC/yr)

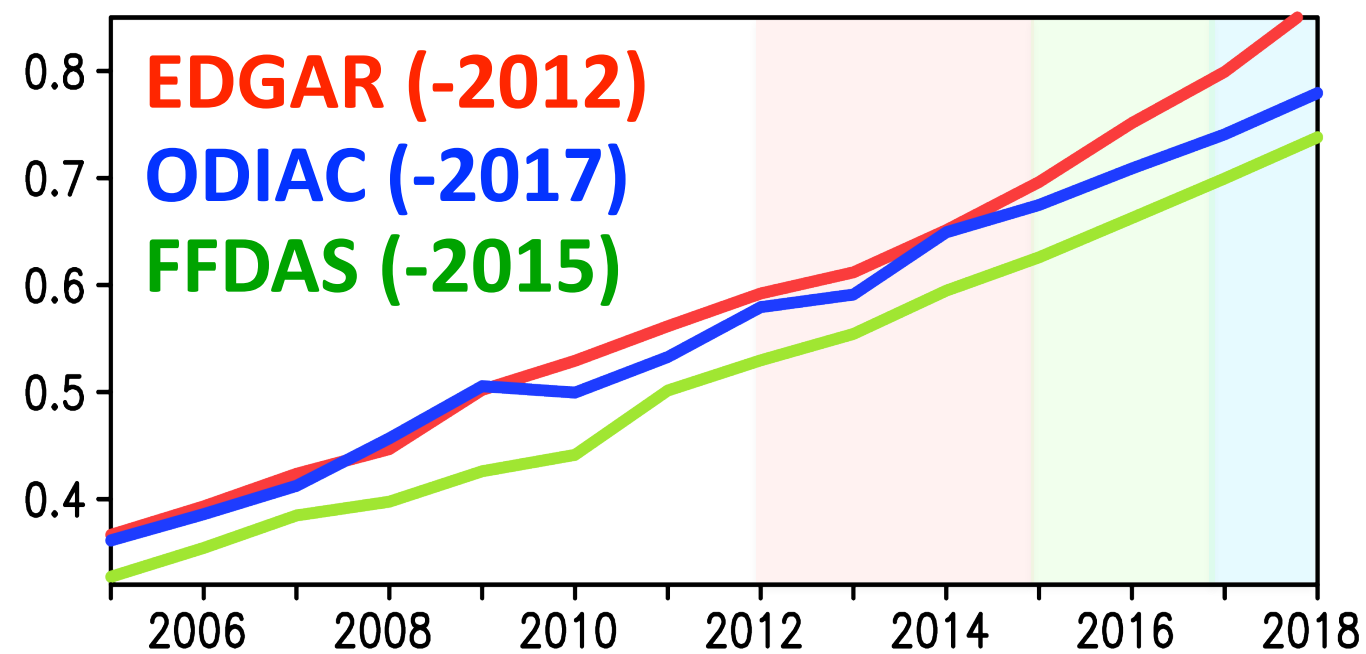
Eastern China



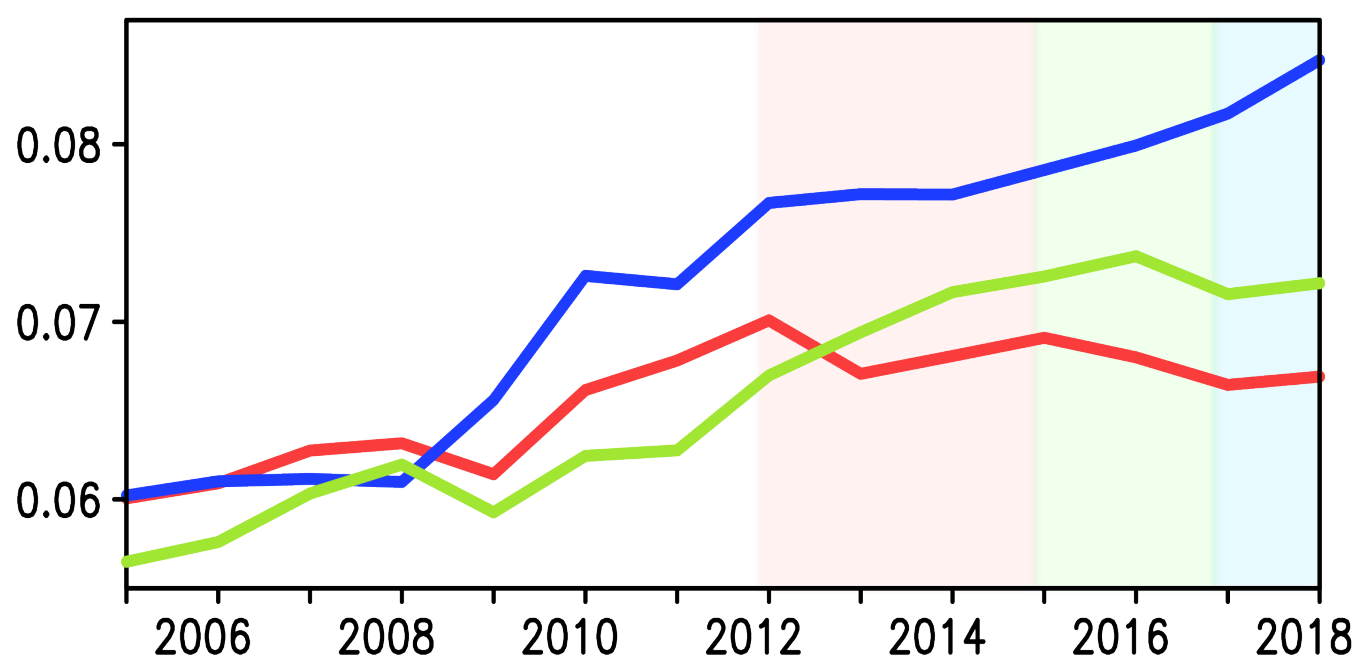
Eastern US



India



Southeast Asia



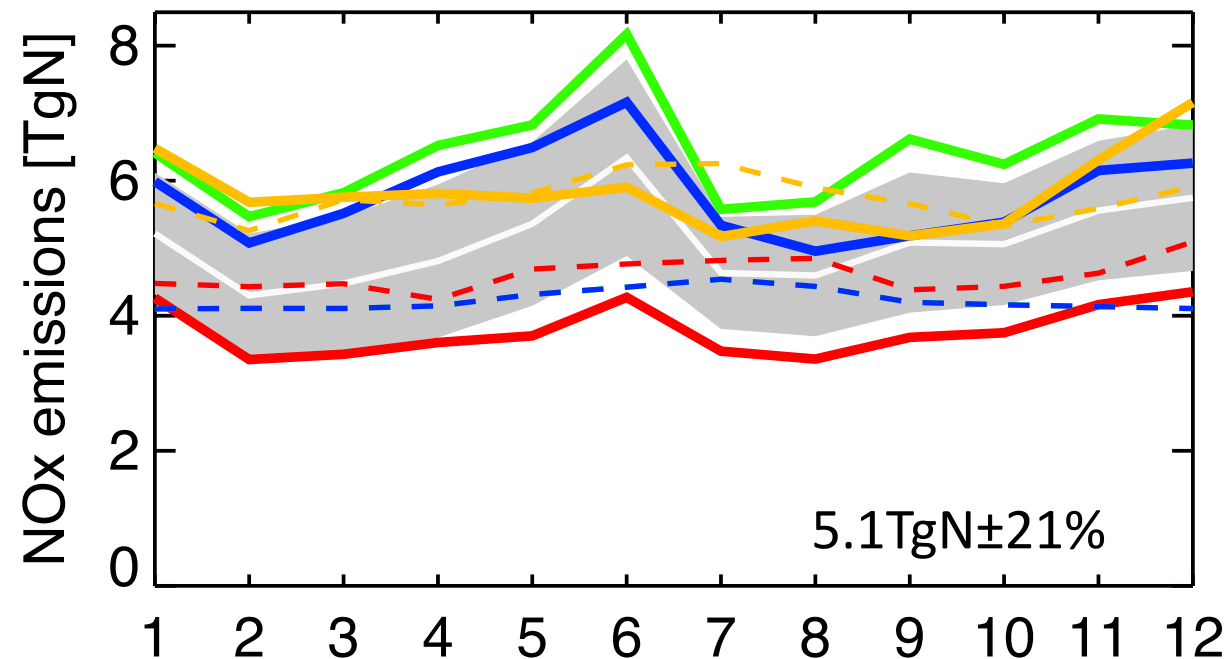
NO_x emission uncertainty



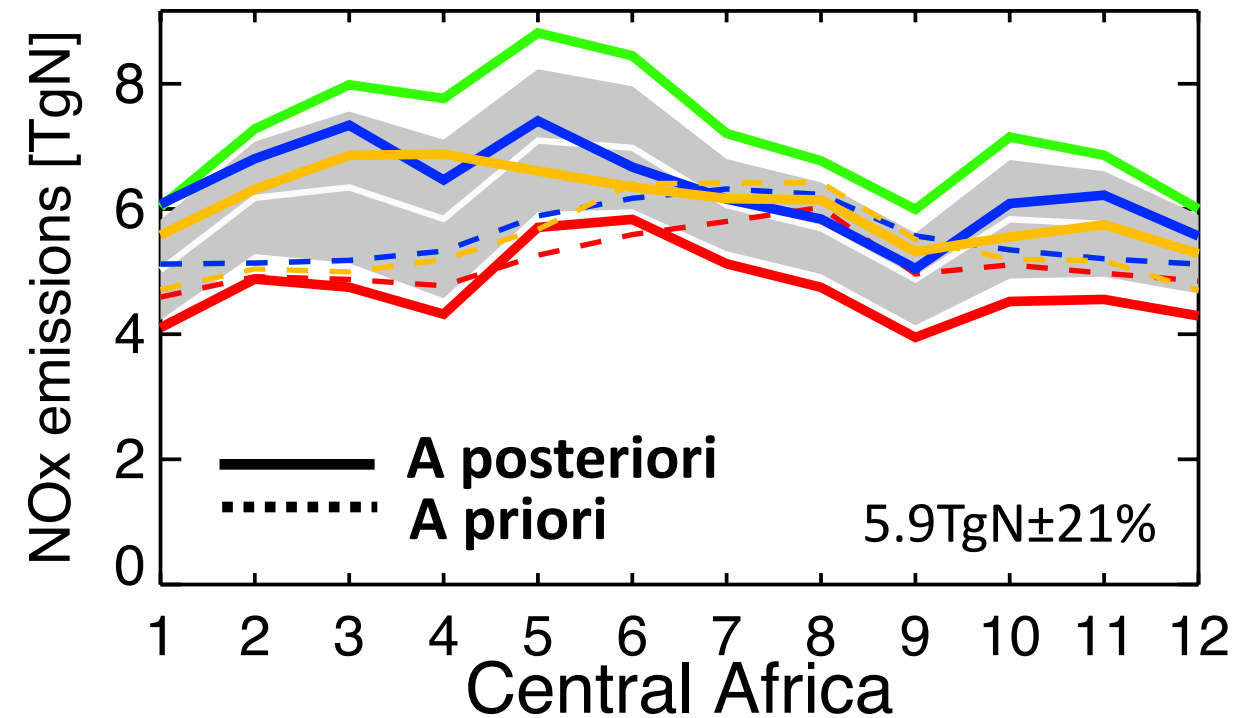
CO₂ flux uncertainty

MOMO-Chem : Multi-mOdel, Multi-cOnstituent CHEMical DA

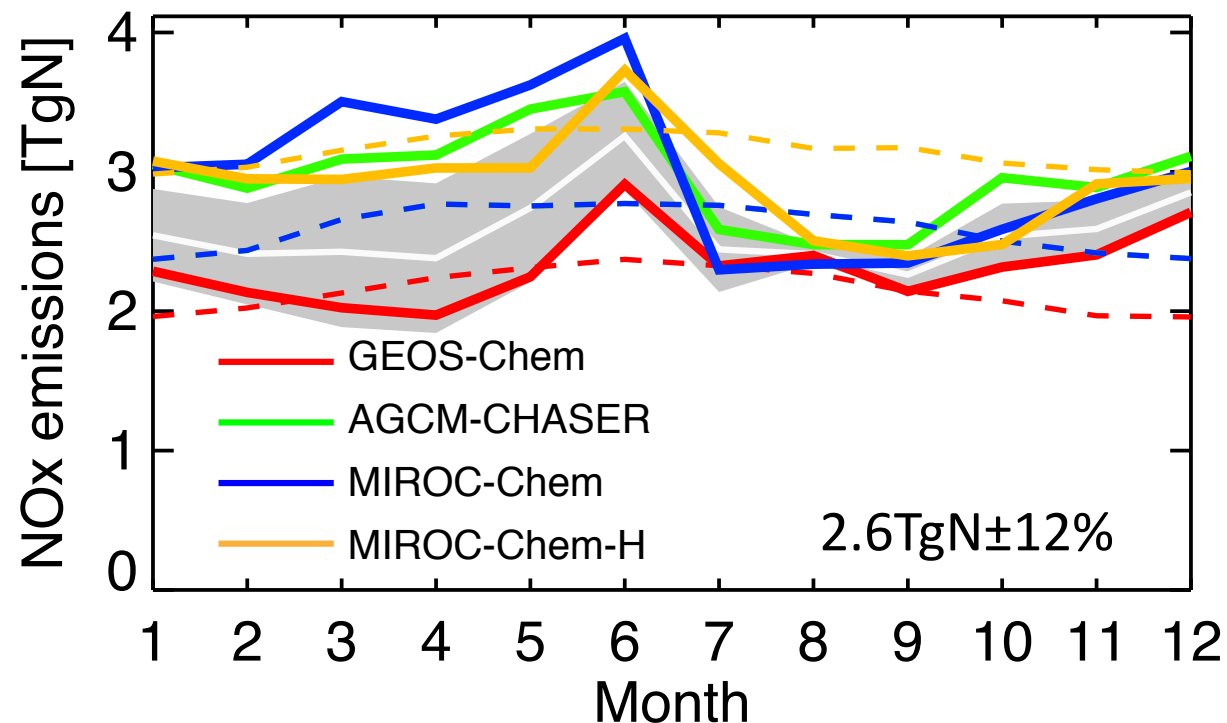
East China



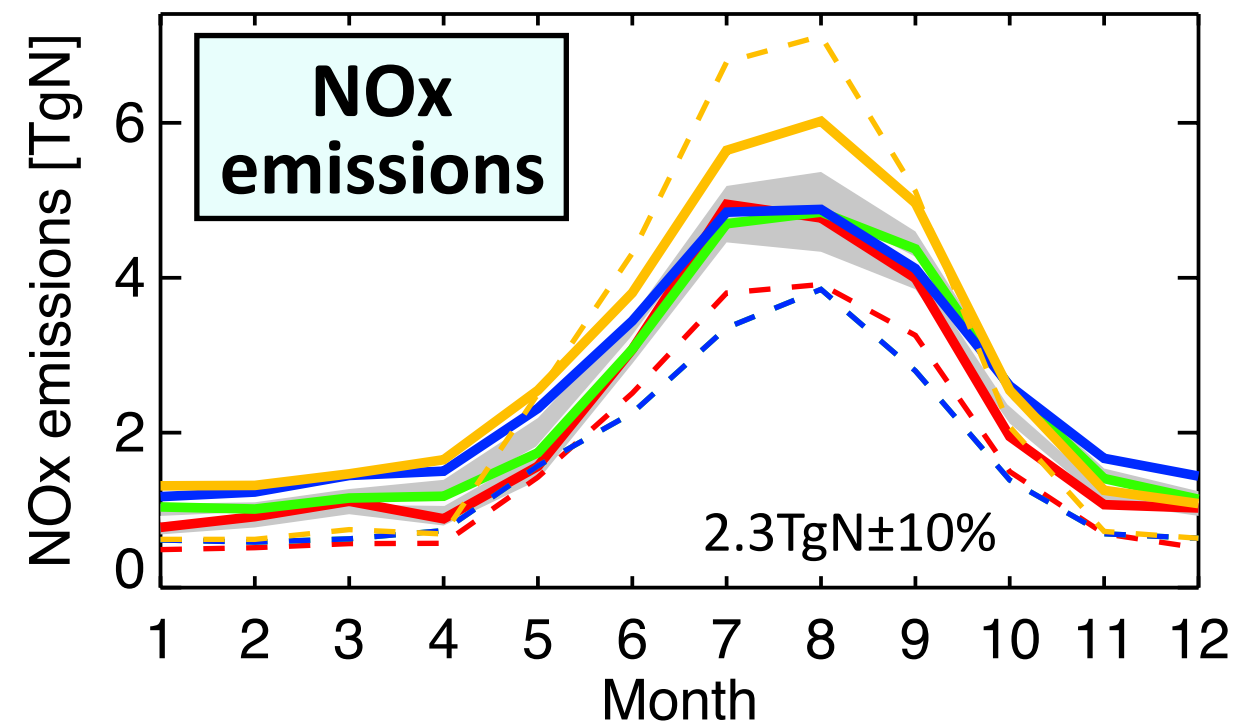
USA



India

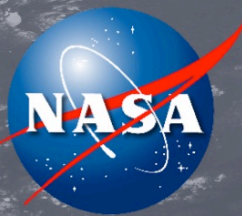


Central Africa



Multi-model SD: 13–31% for industrialized areas and 4–21% for BB areas

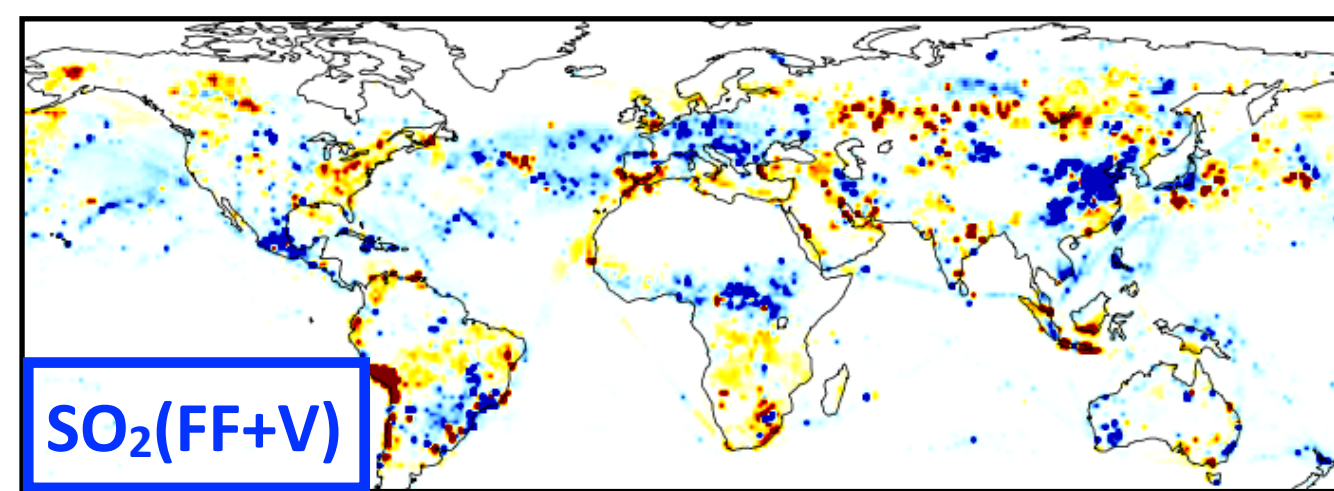
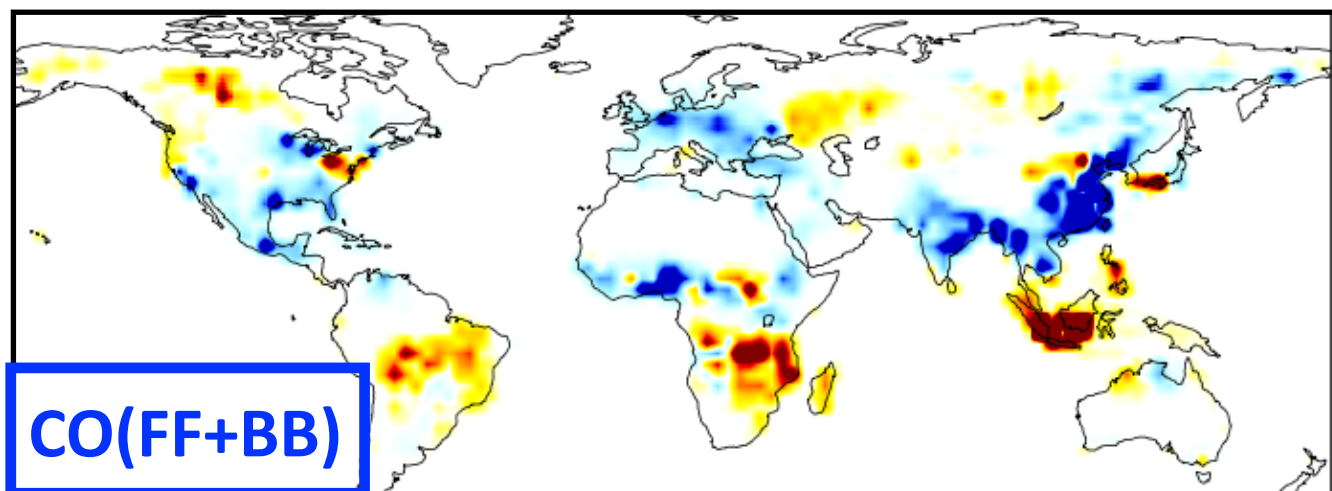
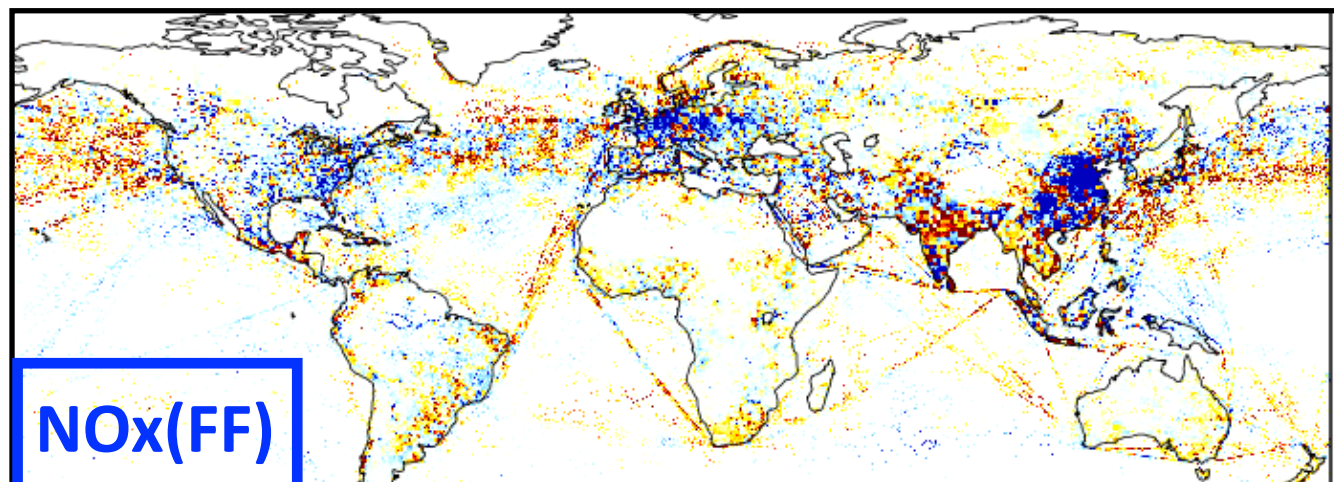
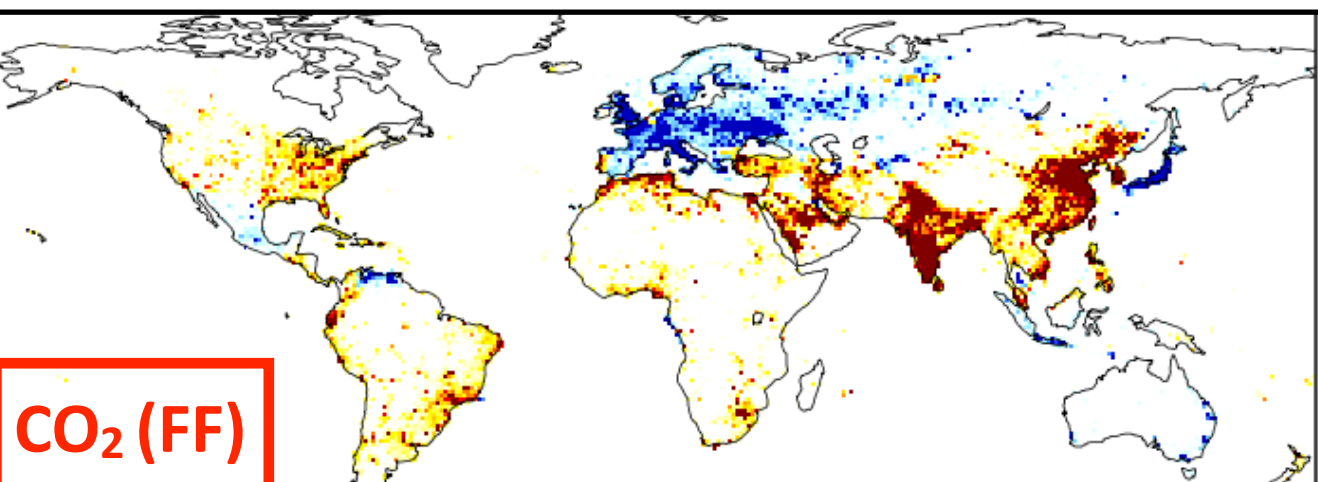
Miyazaki et al., to be submitted



Multi-species constraints on FF CO₂ flux

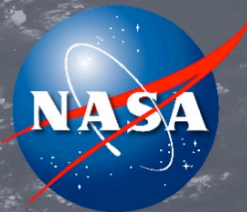
Different aspects of the combustion technology are expected to affect those emissions

- **NO_x**: strongly depend on the temperature of combustion (more NO_x at high T)
- **CO**: can be regarded as a measure of the incompleteness of combustion processes
- **SO₂**: Linked to the burning of fossil fuels. Also, strong emissions from volcanic eruptions.



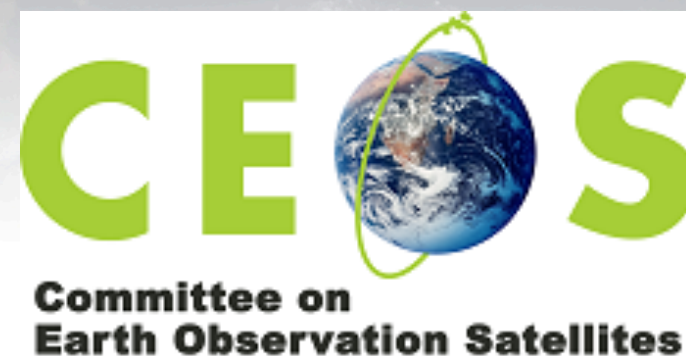
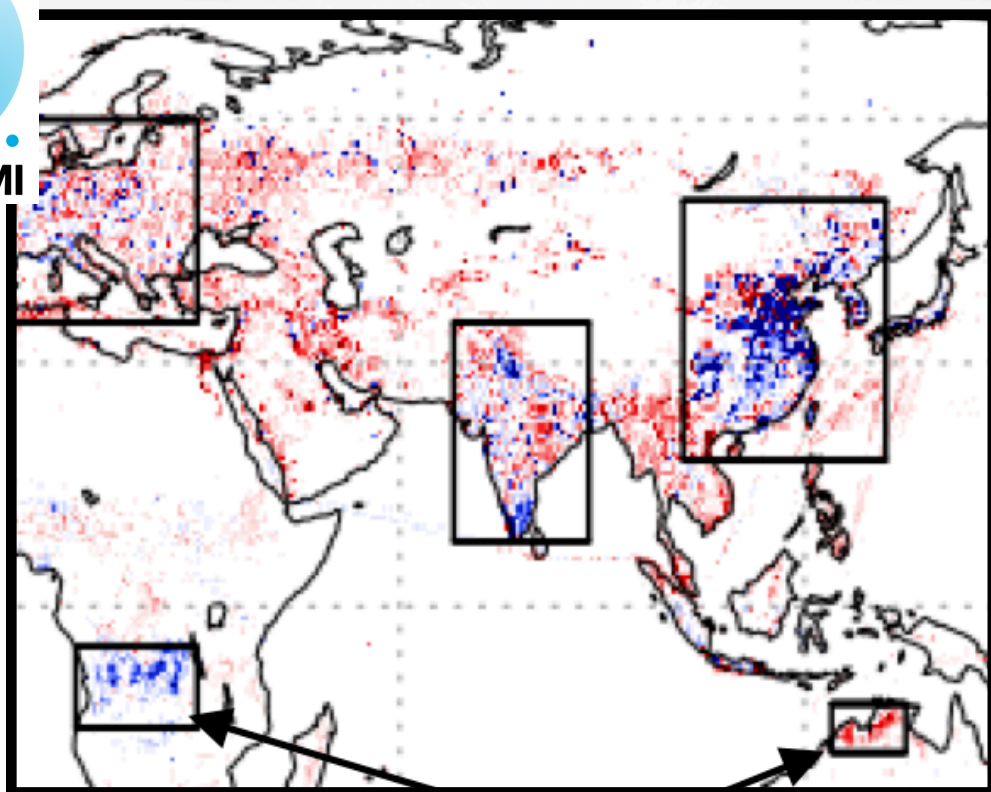
2011-2015
trends

Multi-species regressions using chemical reanalysis products will provide comprehensive constraints on FF CO₂ fluxes and to improve bottom-up inventories



High-resolution multi-species joint emission analysis

Global TROPOMI NO₂ DA at 0.56 deg resolution
(Sekiya et al., poster)

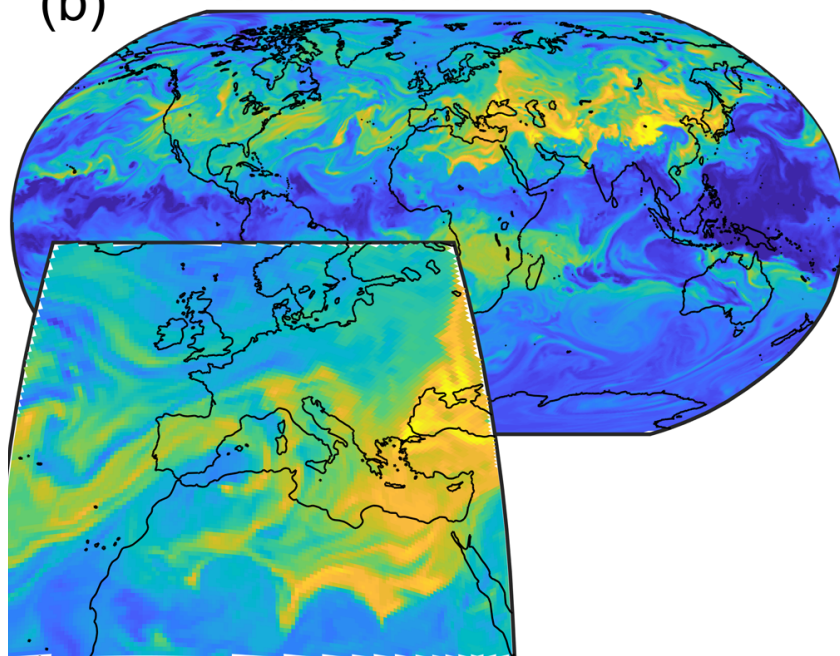


Joint emission estimation
Joint DA analysis

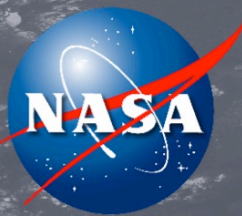


Bottom-up inventories

(b) C180 ozone at 4 km



GCHP-EnKF
developments
at JPL



Conclusions

- FF CO₂ fluxes can be predicted for the most recent years based on Kalman filter trajectories of emission ratios, by combining **bottom-up GHG inventories** with **top-down estimate of proxy species from chemical reanalysis**, which **extend GHG inventories**.
- **The multi-GHG inventories** and **multi-model chemical reanalyses (MOMO-Chem)** provide integrated information on GHG/AQ variations and their uncertainty.
- The obtained long-term changes in emission ratios could suggest developments of multi-species bottom-up inventories, such as REAS and EDGAR.

Future works:

- Multi-species constraints and AQ-GHG joint emission optimizations through high-resolution DA of the existing/future satellites
- Emergent constraints on the chemistry-climate system and carbon cycle

